



Large scale plane wave pseudopotential density functional calculations on GPU clusters

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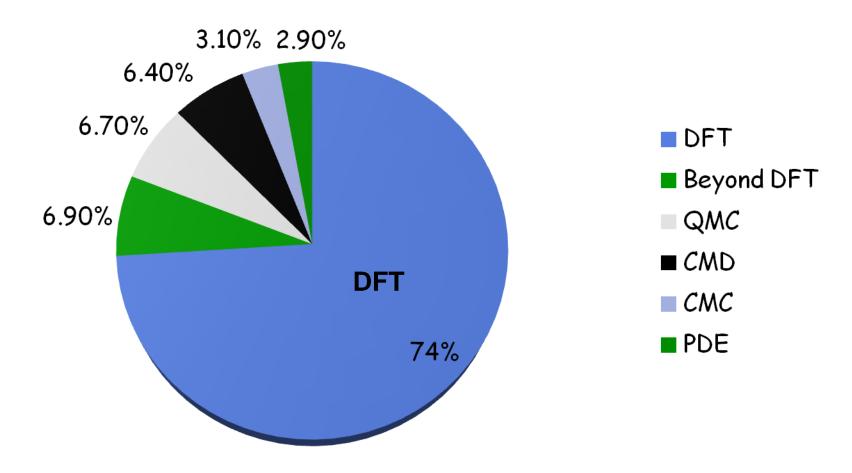
National Basic Research Program of China NSF of China Science & Technology Commission of Shanghai Office of Science, BES, DOE, USA



A profile for material science simulation



A survery of computational material science algorithm in NERSC community (2007)





What is the remaining challenge for DFT calculations?

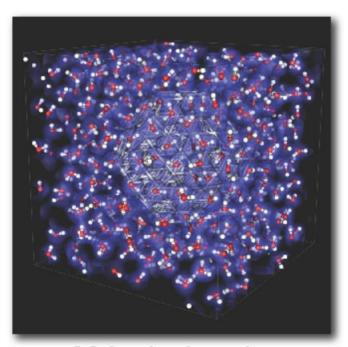


- ❖ 100 to 1000 atoms
- ❖ Ab initio MD for a few ns
- massive configuration space search for structures

State-of-the-art: 1-2 min per MD step (so can only calculate a few ps, But want: ns!)

For >>1000 atoms, linear scaling method

Nanocatalysis: Pt



Molecular dynamics Pt₂₀₁+427H₂O 1482 atoms

P. Kent, ORNL M. Neurock, U. Virginia

Sweet spot: a few hundreds to a few thousand atoms need faster speed

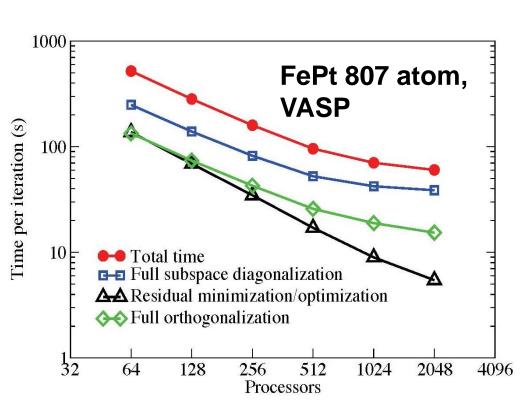


Plane Wave Pseudopotential DFT codes



- ❖ They are the most widely used, and most mature codes
- ❖ There are about a dozen of them: VASP, CASTEP, CPMD, ABINIT, PWSCF, DACAPO, SOCORRO, DFT++, PARATEC, DOD-PW, CP2K, SPHINX, QBOX, PEtot
- ❖ But the CPU codes often do not scale (e.g., 1000 atom system might scale to a few thousand cores)
- **❖** A few minutes per MD step

Idea: use GPU to speed up the absolute speed



P. Kent, ORNL



The computational cost of DFT method



$$\left[-\frac{1}{2}\nabla^2 + V_{tot}(r)\right]\psi_i(r) = \varepsilon_i\psi_i(r)$$

- ◆ If the size of the system is *N*:
- N coefficients to describe one wavefunction $\psi_i(r)$
- i = 1,..., M wavefunctions $\psi_i(r)$, M is proportional to N.
- Orthogonalization: $\int \psi_i(r) \psi_j^*(r) d^3 r$, M^2 wave function pairs, each with N coefficients: N^*M^2 , i.e N^3 scaling.

The repeated calculation of these orthogonal wave functions make the computation expensive, $O(N^3)$.



PEtot code



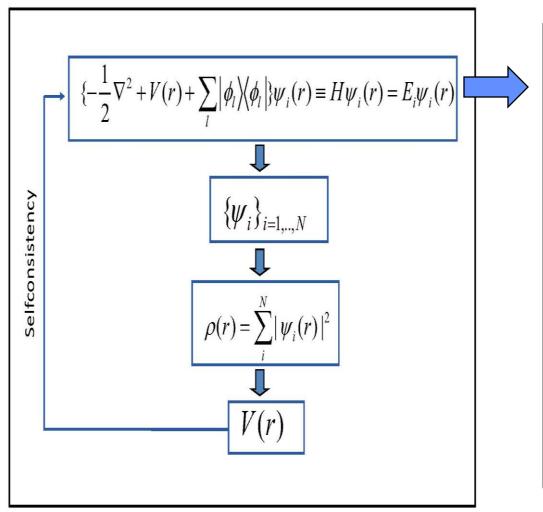
- ❖ Developed in Lawrence Berkeley National Lab
- Free: https//hpcrd.lbl.gov/~linwang/PEtot/PEtot.html
- ❖ Has three levels of parallelization: G-space, state index, k-point
- Uses norm conserving pseudopotential and ultra-soft psd.
- Use parallel FFT (by Andrew Canning)
- **❖** Can calculate 10,000 states on a few thousand processors



The flow chart of the DFT method (PEtot code)



The overall flow chart of SCF iterations



The conjugate-gradient (CG) to solve the Schrodinger's eq (98% of the total time)

$$h(i,j) = \left\langle \psi_i \middle| H \middle| \psi_j \right\rangle \qquad \text{Sub_diag, *}$$

$$P_i = H \psi_i - \varepsilon_i \psi_i \qquad \text{Hpsi, *}$$

$$P_i = A(P_i - \frac{\lambda_i}{\lambda_i^o} P_i^o) \qquad \text{Precond. CG step}$$

$$P_i = P_i - \sum_{j=1,i} \left\langle P_i \middle| \psi_j \right\rangle \qquad \text{Projection, *}$$

$$\psi_i = \psi_i \cos \theta_i + P_i \sin \theta_i \qquad \text{Line minimiz.}$$

$$\psi_i = \psi_i - \sum_{j < i} \left\langle \psi_i \middle| \psi_j \right\rangle \qquad \text{Orth., *}$$

$$h(i,j) = \left\langle \psi_i \middle| H \middle| \psi_j \right\rangle \qquad \text{Sub_diag, *}$$



The kernels in the H*ψ (Hpsi)

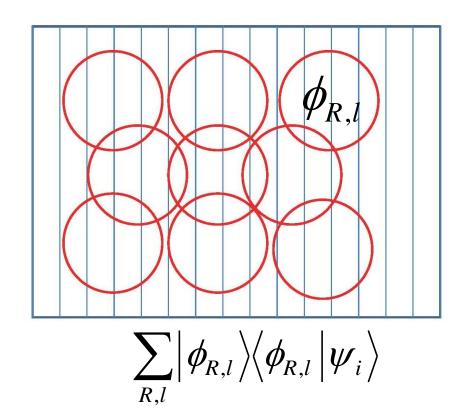


$$\left\{-\frac{1}{2}\nabla^2 + V(r) + \sum_{l} \left|\phi_l\right\rangle \left\langle\phi_l\right| \right\} \psi_i(r)$$

FFT (by A. Canning)

P2 P1 PO X

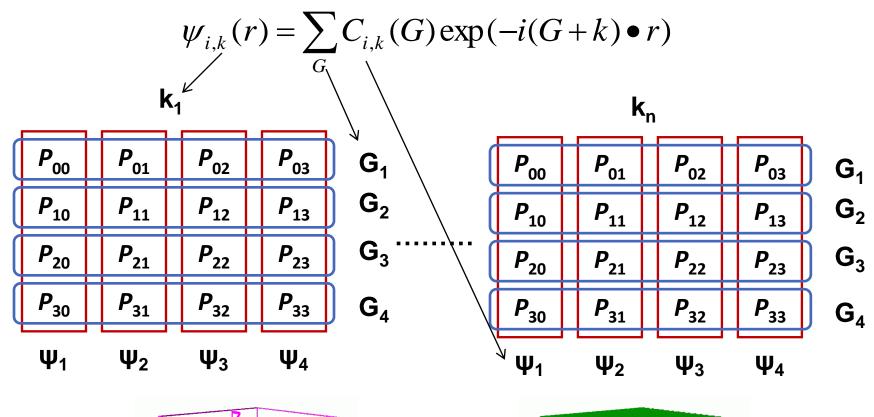
Real sace Nonlocal pseudopotential



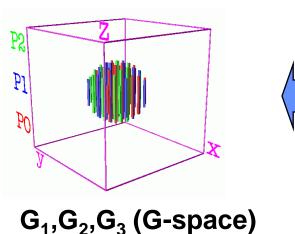


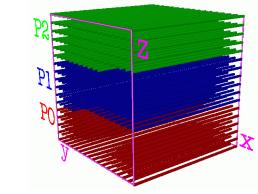
Parallelization scheme for a CPU code





Parallel FFT (each CPU has many 1D FFTs)





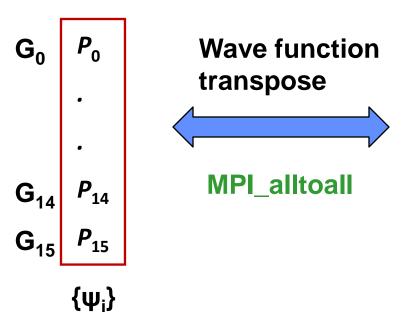
Real space



GPU hybrid parallelization



G-parallel



$\langle \psi_i | \psi_j \rangle$

Diag rotation

CUBLAS MPI_allreduce

Index parallel

$$P_0 cdot cdot cdot cdot ext{P}_{14} ext{ P}_{15} ext{ } \{G\}$$
 $\Psi_0 ext{ } \Psi_{14} ext{ } \Psi_{15} ext{ }$

Hpsi FFT nonlocal

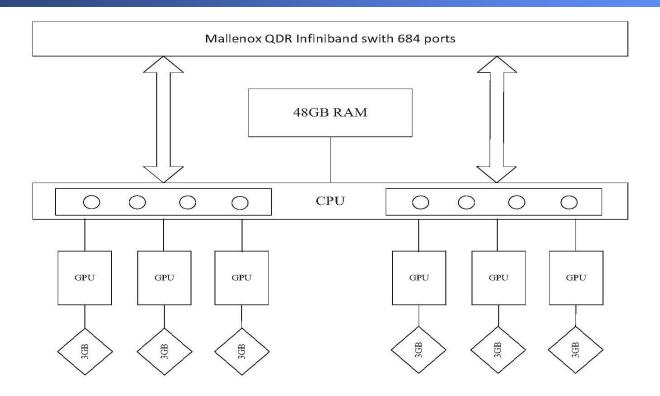
CUFFT

- The FFT is within a single GPU (no parallel FFT)
- memory limitation to the size: a few thousand atoms



A single node in the CPU/GPU machine (IPE)





CPU: Xeon 5520 quad-core CPU 9 Gflops/core (2.2 GHz) 6 GB memory/core

GPU: Nvidia Fermi C2050 GPU card 448 stream processors/card 515 Gflops/card (double precision) 3 GB memory/card

Multiple GPU cards in one node (Institute of Processing Engineering, CAS)

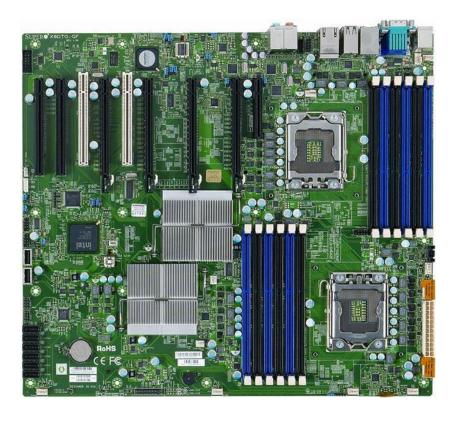
Strategy: one CPU core controls one GPU card, CPU/GPU unit



Another example of multiple GPU per node machine







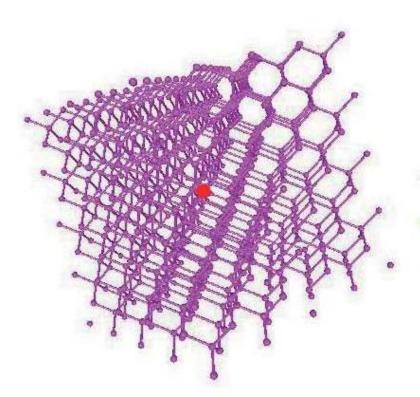
- **❖ NEWTON**, offered by Electronics Nexus
- ❖ 8 CPU cores (Intel)
- ❖ 8 GPU cards (Nvidia)
- **❖** Start from \$2,199!

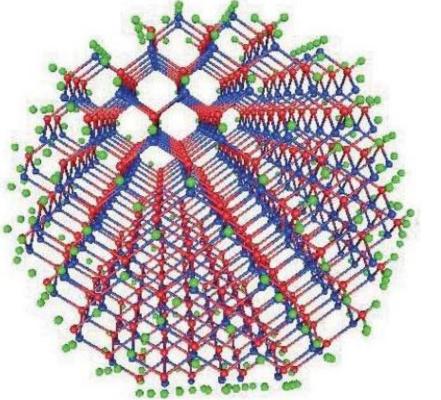


The testing systems



GaAs:N (512 atoms) 2048 electrons 128³ FFT grid 40 Ryd Ecut 3.3 x10⁵ PW coeff CdSe quantum dot (933 atoms) 2832 electrons 256³ FFT grid 30 Ryd Ecut 1.1x10⁶ PW coeff







GPU coding (easy to use CUBLAS)



CALL zgemm('c','n',mx, mx,ng_n,one,A,mg,B,mg, zero,SS, mx)

CPU code



```
stat = cublas_alloc(mg*mx, 16, cu_A) ! Alloc CUDA memory
stat = cublas_alloc(mx*mx, 16, cu_SS)
stat = cublas_alloc(mg*mx, 16, cu_B)
call cublas_set_matrix (mg, mx, 16, A, mg, cu_A, mg) ! Copy matrix to GPU
call cublas_set_matrix (mg, mx, 16, B, mg, cu_B, mg)
call cublas_zgemm('c','n',mx,mx,ng_n,one,cu_A,mg, cu_B,mg, zero,cu_SS,mx) ! Cublas call
call cublas_get_matrix (mx, mx, 16, cu_SS, mx, SS, mx) ! Get matrix to CPU
call cublas_free(cu_A)
call cublas_free(cu_B)
call cublas_free(cu_SS) ! Free CUDA memory
```

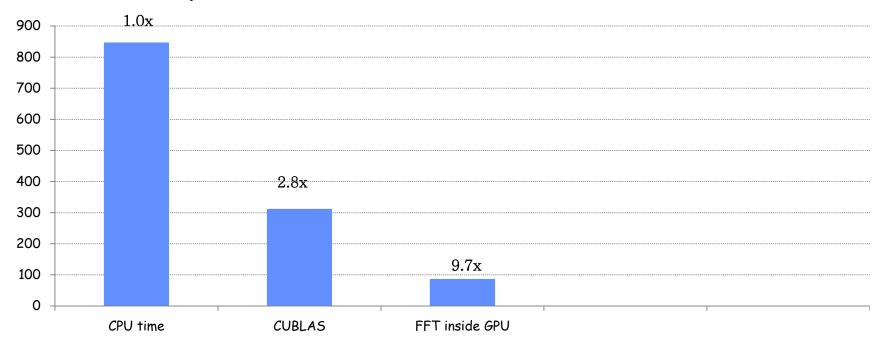
GPU code



Different steps of speeding up to go to GPU



Computation Time for CG_AB (16 CPU/GPU units)





The results



Computing units	16	32	64	128	256	256
systems	512-GaAs	512-GaAs	512-GaAs	512-GaAs	512-GaAs	933-CdSe
PEtot (CPU)	842	450	255	152	104	495
PEtot (GPU)	87	49	27	23	17	56
Speed-up (PEtot)	9.7x	9.2x	9.4x	7x	6.1x	8.8x
Total flops (Tflops)	0.59	1.05	1.91	2.24	3.03	5.92
Efficiency	7.1%	6.3%	5.7%	3.3%	2.3%	4.4%

Computing unit: one CPU core/ one GPU card

Times: in seconds

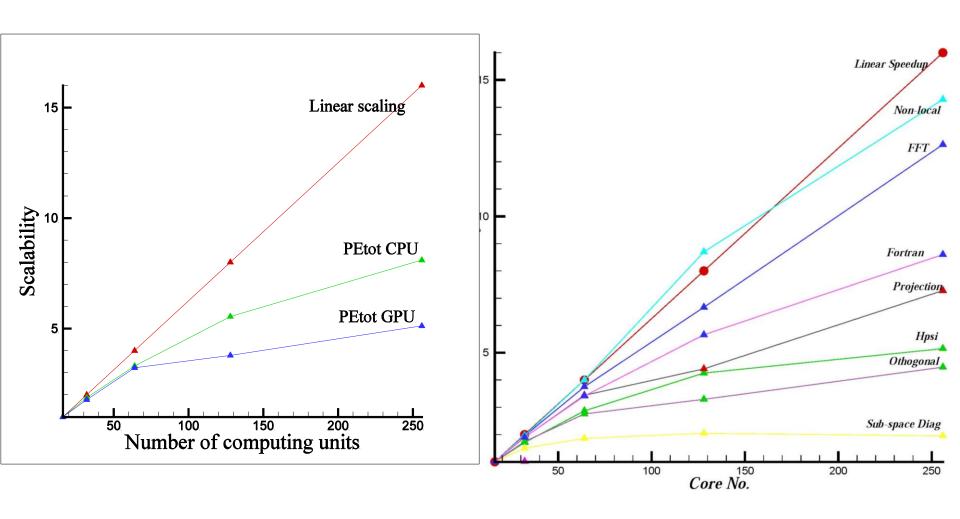
4 line min steps in CG_AB

Only the CG_AB times are reported



The processor scalings

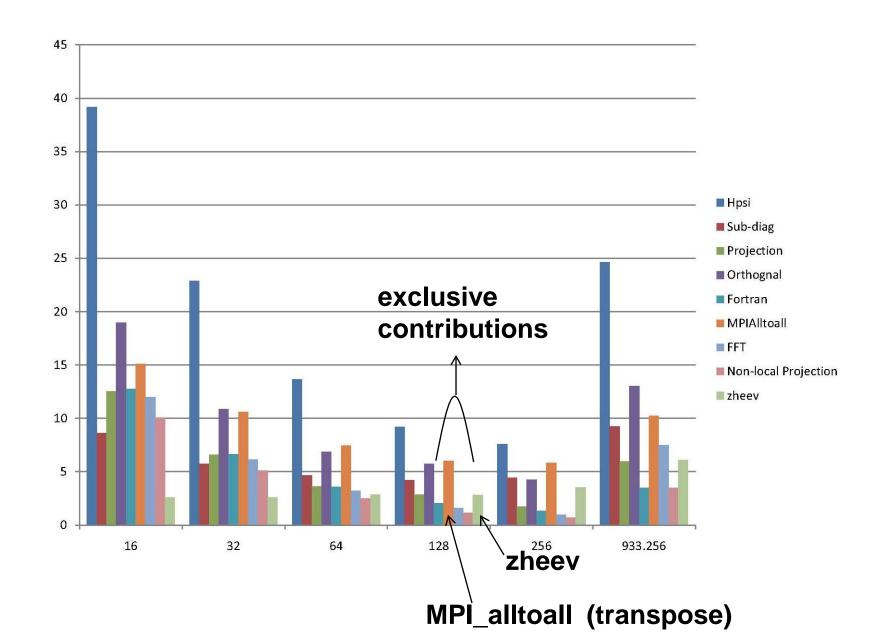






The total computational times for different kernels







The remaining problems & solutions



❖ The MPI_alltoall (for transpose) takes time

For P=Hψ-εψ and H*P, reduce the double precision to 4 byte number, hence reduce the MPI_alltoall

❖The matrix diagonalization routines take time

Using new CPU and GPU routines for diagonalizations

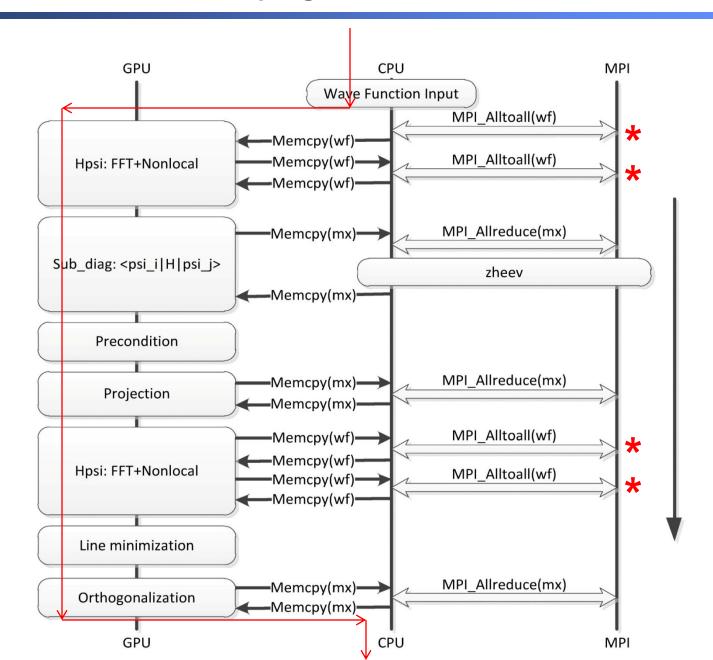
❖ The CPU-GPU wave function data copies take time

Move all the computations to GPU, reduce CPU-GPU data copy



The new program flow chart

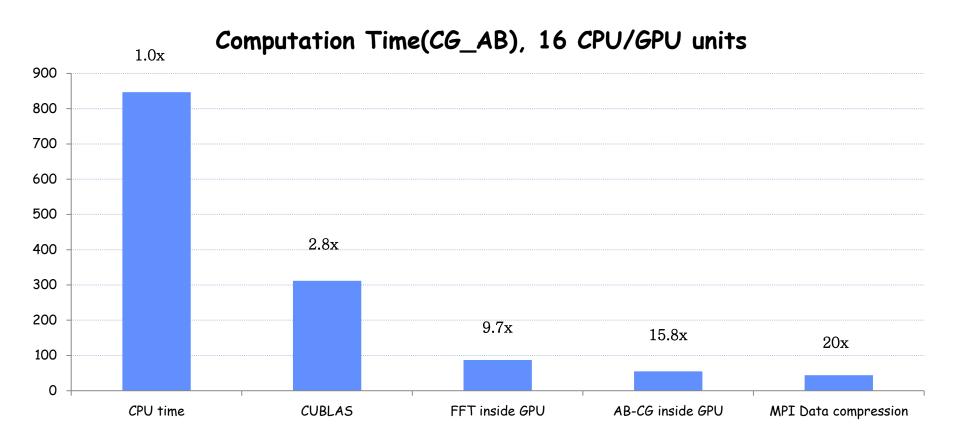






Different steps of speeding up to go to GPU







CONCLUSIONS



- ❖ It is possible to use GPU to speed up PW Pseudopotential DFT code by x20.
- ❖ Need to change the parallelization scheme, and introduce new algorithm.
- Hpsi and FFT are done within one GPU
- ❖ Want as many GPU per node as possible, CPU not used
- Want large GPU global memory (one whole wave function will be stored in one GPU)
- **❖** Want faster MPI_alltoall, MPI_allreduce
- Want faster GPU multi-processor lib